

Homework 3: Reconstruct

Course: CS 221 Spring 2019

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Setup: n-gram Language Models and Uniform-Cost Search

Our algorithm will base segmentation and insertion decisions on the cost of processed text according to a language model. A language model is some function of the processed text that captures its fluency.

A very common language model in NLP is an n-gram sequence model. This is a function that, given n consecutive words, gives a cost based on to the negative log likelihood that the n -th word appears just after the first $n-1$. The cost will always be positive, and lower costs indicate better fluency. As a simple example: in a case where $n = 2$ and c is our n-gram cost function, $c(\text{big}, \text{fish})$ would be low, but $c(\text{fish}, \text{fish})$ would be fairly high.

Furthermore, these costs are additive; for a unigram model $u(n = 1)$, the cost assigned to $[w_1, w_2, w_3, w_4]$ is

$$u(w_1) + u(w_2) + u(w_3) + u(w_4).$$

For a bigram model $b(n = 2)$, the cost is

$$b(w_0, w_1) + b(w_1, w_2) + b(w_2, w_3) + b(w_3, w_4)$$

where w_0 is -BEGIN-, a special token that denotes the beginning of the sentence.

We have estimated u and b based on the statistics of n -grams in text. Note that any words not in the corpus are automatically assigned a high cost, so you do not have to worry about this part.

A note on low-level efficiency and expectations: this assignment was designed considering input sequences of length no greater than roughly 200 (characters, or list items, depending on the task). Of course, it's great if programs tractably manage larger inputs, but it isn't expected that such inputs not lead to inefficiency due to overwhelming state space growth.

Problem 1: Word Segmentation

In word segmentation, you are given as input a string of alphabetical characters ($[a-z]$) without whitespace, and your goal is to insert spaces into this string such that the result is the most fluent according to the language model.

- (a) Consider the following greedy algorithm: Begin at the front of the string. Find the ending position for the next word that minimizes the language model cost. Repeat, beginning at the end of this chosen segment.

Show that this greedy search is suboptimal. In particular, provide an example input string on which the greedy approach would fail to find the lowest-cost segmentation of the input.

In creating this example, you are free to design the n-gram cost function (both the choice of n and the cost of any n-gram sequences) but costs must be positive and lower cost should indicate better fluency. Note that the cost function doesn't need to be explicitly defined. You can just point out the relative cost of different word sequences that are relevant to the example you provide. And your example should be based on a realistic English word sequence don't simply use abstract symbols with designated costs.

One example would be the input string "basketballismyfavoritesport". If "basket" has a lower cost than "basketball", the result might be "basket ball is my favorite sport" instead of "basketball is my favorite sport". The algorithm will not look ahead to notice that "basketball is" has a lower cost than "basket ball is".

- (b) coding